

# Automated Machine Learning (AutoML)

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# Motivation

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1. Machine learning is very **successful**
2. To build a traditional ML pipeline:
  - Domain experts with longstanding experience
  - Specialized data preprocessing
  - Domain-driven meaningful feature engineering
  - Picking right models
  - Hyper-parameter tuning
  - ... ..

# AutoML Vision

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## For Non-Experts

AutoML allows non-experts to make use of machine learning models and techniques without requiring to become an expert in this field first

[https://en.wikipedia.org/wiki/Automated\\_machine\\_learning](https://en.wikipedia.org/wiki/Automated_machine_learning)

## For Data Scientists

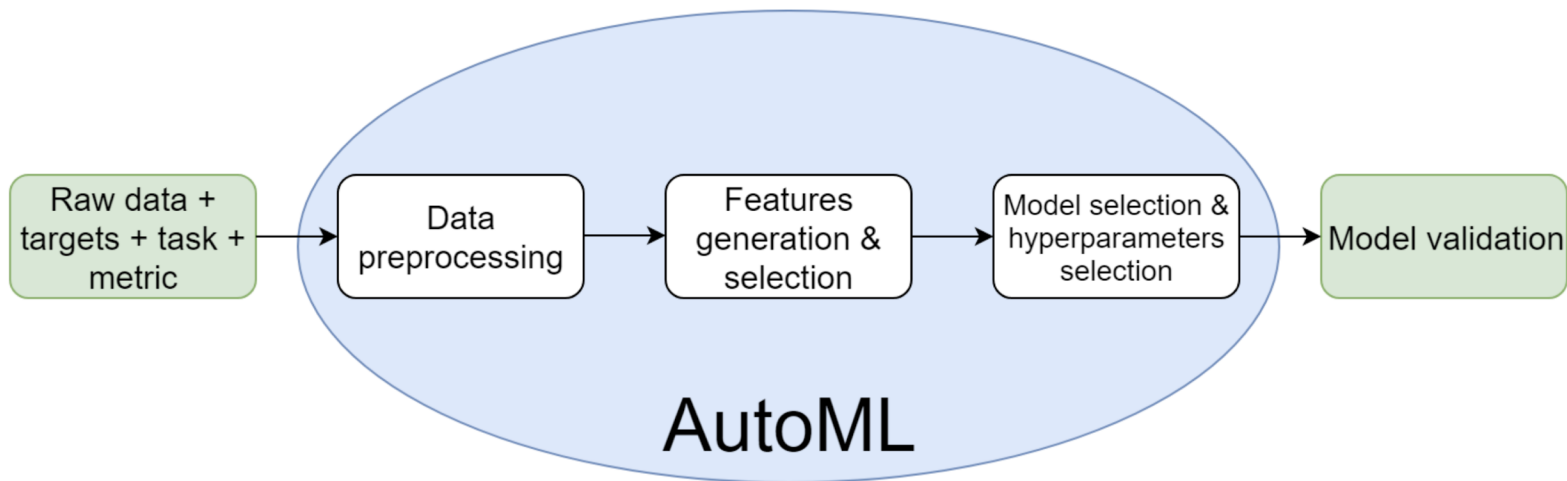
AutoML aims to augment, rather than automate, the work and work practices of heterogeneous teams that work in data science.

[Wang, Dakuo, et al. "Human-AI Collaboration in Data Science: Exploring Data Scientists' Perceptions of Automated AI." Proceedings of the ACM on Human-Computer Interaction 3.CSCW \(2019\): 1-24.](#)

# What is AutoML?

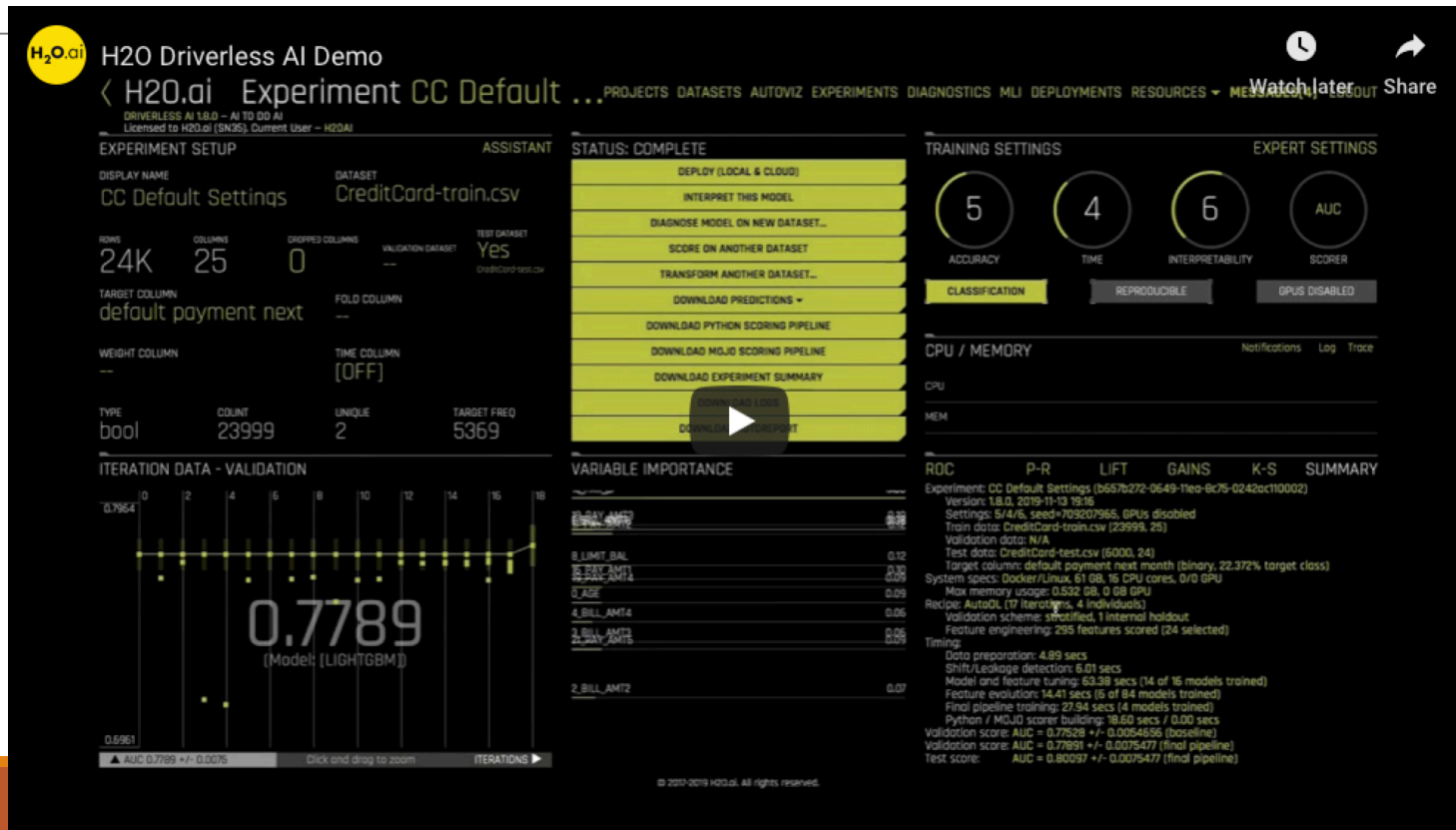
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- ❖ Automate the process of applying machine learning to real-world problems



# H2O Driverless AI Demo

<https://www.youtube.com/watch?v=ZqCoFp3-rGc>



# Outline

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Auto Feature Selection (Lecture 6)

Auto Hyperparameter Tuning (Lecture 6)

**Auto Feature Generation (This Lecture)**

**Neural Architecture Search (This Lecture)**

# Auto Feature Generation

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# Motivation

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- ❖ The model performance is heavily dependent on quality of features in dataset
- ❖ It's time-consuming for domain experts to generate enough useful features





# Feature Generation

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- ❖ Unary operators (applied on a single feature)
  - Discretize or normalize numerical features
  - Apply rule-based expansions of dates
  - Mathematical operators (e.g., Log Function)
- ❖ Higher-order operators (applied on 2+ features)
  - Basic arithmetic operations (e.g.,  $+$ ,  $-$ ,  $\times$ ,  $\div$ )
  - Group-by Aggregation (e.g., GroupByThenMax, GroupByThenMin)

# Featuretools



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- ❖ An open source library for performing automated feature engineering
  - ❖ Design to fast-forward feature generation across **multi-relational** tables

# Concepts

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- ❖ **Entity** is the relational tables
- ❖ An **EntitySet** is a collection of entities and the relationships between them
- ❖ **Feature Primitives**
  - ❖ Unary Operator: transformation (e.g., MONTH)
  - ❖ High-order Operator: Group-by Aggregation (e.g., GroupByThenSUM)

# Entity sets

## Customer

Customer_id	Birthdate	MONTH(Birthdate)	SUM(Product.Price)
1	1995-09-28	9	\$500
2	1980-01-01	1	...
3	1999-02-02	2	...
...	...	...	...

## Product

Product_id	Customer_id	Name	Price
1	1	Banana	\$100
2	1	Banana	\$100
3	1	Orange	\$300
4	2	Apple	\$50
...	...	...	...

GroupBy  
ThenSUM:

Unary Operator:  
MONTH

Feature  
Primitives

# Outline

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Auto Feature Selection (Lecture 5)

Auto Hyperparameter Tuning (Lecture 5)

**Auto Feature Generation (This Lecture)**

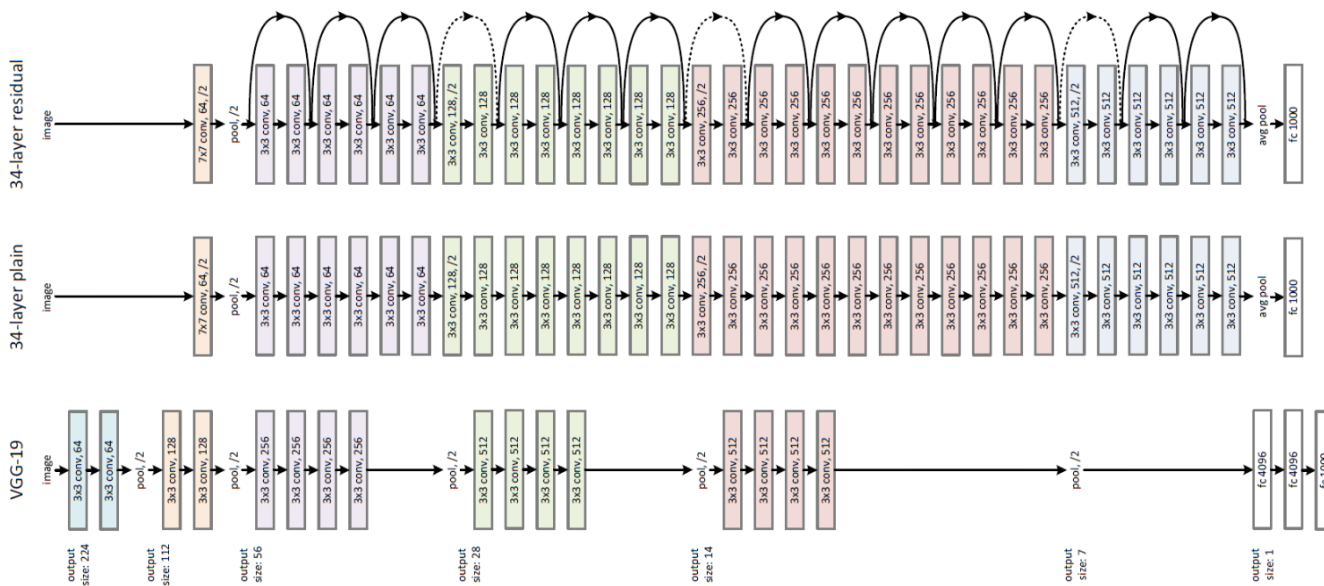
**Neural Architecture Search (This Lecture)**

# Neural Architecture Search (NAS)

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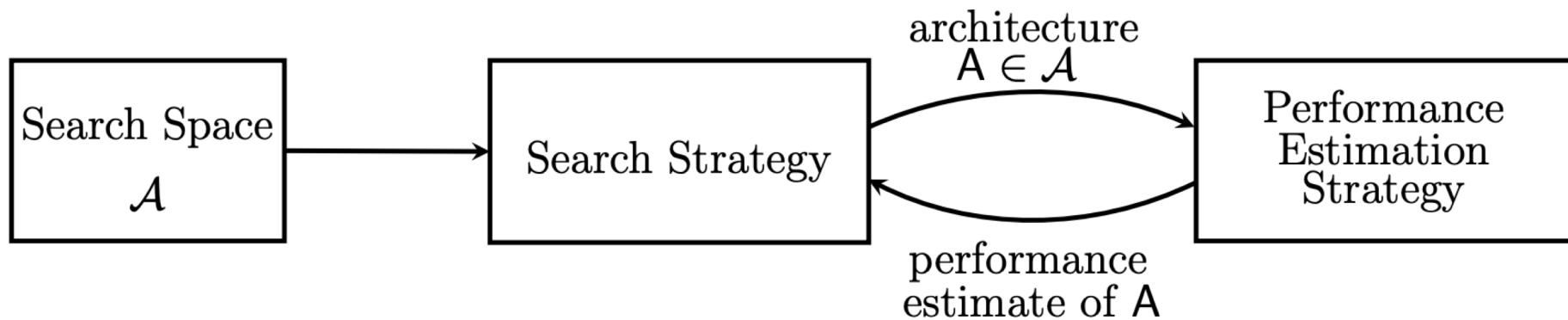
# Motivation

How can someone come out with such an architecture?



# Neural Architecture Search: Big Picture

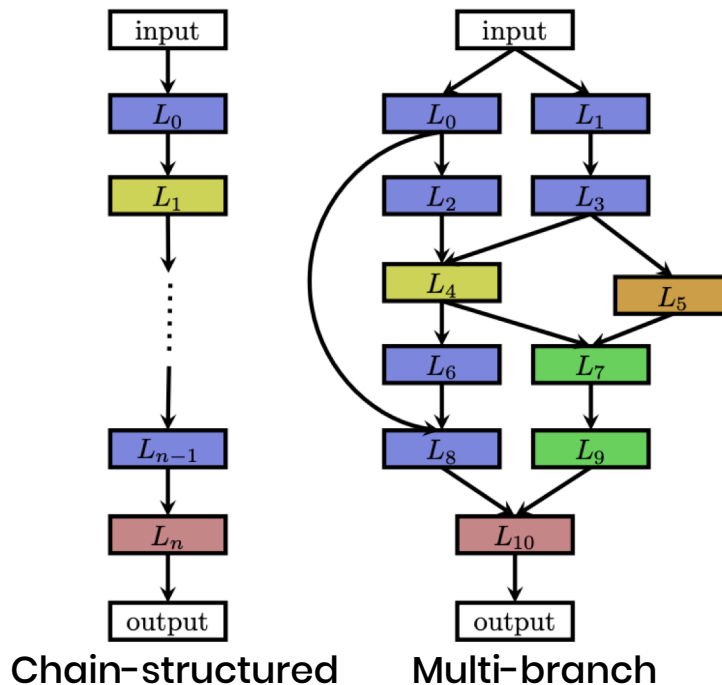
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# Search Space

- ❖ Define which neural architectures a NAS approach might discover in principle
- ❖ May have human bias → prevent finding novel architectural building blocks



# Search Strategy

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## ❖ Basic Idea

- Explore search space (often exponentially large or even unbounded)

## ❖ Methods

- Random Search
- Evolutionary Methods [Angeline et al., 1994]
- Bayesian Optimization [Bergstra et al., 2013]
- Reinforcement Learning [Baker et al., 2017]
- .....

# Performance Estimation Strategy

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## ❖ Basic Idea

- The process of estimating predictive performance

## ❖ Methods

- Simplest option: perform a training and validation of the architecture on data
- Initialize weights of novel architecture based on weights of other architectures have been trained before
- Using learning curve extrapolation [Swersky et al., 2014]
- .....

# Summary

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**What is AutoML and why we need it?**

**How AutoML works?**

- Auto Feature Selection (Lecture 5)
- Auto Hyperparameter Tuning (Lecture 5)
- Auto Feature Generation (This Lecture)
- Neural Architecture Search (This Lecture)